### Lecture 16 Thur. 3.10.2016

Introduction to coupled oscillation and eigenmodes (Ch. 3-4 of Unit 2)

Review of 1D FDHO (Forced-Damped-Harmonic Oscillator) response

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms

2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag)

Eigensolutions by matrix-algebra with example  $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues  $\Rightarrow$  eigenvectors) Operator orthonormality and Completeness (Idempotent means:  $\mathbf{P} \cdot \mathbf{P} = \mathbf{P}$ ) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors



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Thursday, March 10, 2016



 $\frac{\text{OscillIt Web Simulation}}{\text{Lorentz Response (}\Gamma=0.2)}$ 



<u>OscillIt Web Simulation</u> <u>Lorentz Response ( $\Gamma$ =0.2)</u>



 $\frac{\text{OscillIt Web Simulation}}{\text{Lorentz Response (}\Gamma=0.2)}$ 



OscillIt Web Simulation Lorentz Response (T=0.2)



<u>OscillIt Web Simulation</u> <u>Lorentz Response ( $\Gamma$ =0.2)</u>

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Fig. 3.3.3 One 2-dimensional coupled oscillator

2D HO kinetic energy  $T(v_1, v_2)$ 

$$T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$$



2D HO kinetic energy T(v<sub>1</sub>, v<sub>2</sub>)  $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$   $2D \text{ HO potential energy } V(x_1, x_2)$  $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$  $= \frac{1}{2}(k_1 + k_{12})x_1^2 - k_{12}x_1x_2 + \frac{1}{2}(k_2 + k_{12})x_2^2$ 

2D harmonic oscillator (2D-HO) equations of motion
Lagrangian and matrix forms



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Lagrange-Newton equations for 2D HO

$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{x}_1} \right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right) x_1 + k_{12} x_2$$
$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{x}_2} \right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12} x_1 - \left(k_2 + k_{12}\right) x_2$$



2D HO kinetic energy  $T(v_1, v_2)$ 2D HO potential energy  $V(x_1, x_2)$  $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$  $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}(x_1 - x_2)^2$  $=\frac{1}{2}(k_1+k_{12})x_1^2-k_{12}x_1x_2+\frac{1}{2}(k_2+k_{12})x_2^2$ Lagrange-Newton equations for 2D HO

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$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{x}_2} \right) = m_2 \ddot{x}_2 = F_2 = -\frac{\partial V}{\partial x_2} = k_{12} x_1 - \left(k_2 + k_{12}\right) x_2$$

2D HO Matrix operator equations

$$\begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$



2D HO kinetic energy  $T(v_1, v_2)$ 2D HO potential energy  $V(x_1, x_2)$  $T = \frac{1}{2}m_1\dot{x}_1^2 + \frac{1}{2}m_2\dot{x}_2^2$  $V = \frac{1}{2}k_1x_1^2 + \frac{1}{2}k_2x_2^2 + \frac{1}{2}k_{12}\left(x_1 - x_2\right)^2$  $=\frac{1}{2}(k_1+k_{12})x_1^2-k_{12}x_1x_2+\frac{1}{2}(k_2+k_{12})x_2^2$ Lagrange-Newton equations for 2D HO

$$\frac{d}{dt} \left( \frac{\partial T}{\partial \dot{x}_1} \right) = m_1 \ddot{x}_1 = F_1 = -\frac{\partial V}{\partial x_1} = -\left(k_1 + k_{12}\right) x_1 + k_{12} x_2$$
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2D HO Matrix operator equations

$$\begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Matrix operator notation:

$$\mathbf{M} \cdot \left| \ddot{\mathbf{x}} \right\rangle = - \mathbf{K} \cdot \left| \mathbf{x} \right\rangle$$



$$T = \frac{1}{2} m_1 \dot{x}_1^2 + \frac{1}{2} m_2 \dot{x}_2^2 \qquad V = \frac{1}{2} (k_1 + k_{12}) x_1^2 - k_{12} x_1 x_2 + \frac{1}{2} (k_2 + k_{12}) x_2^2 = \frac{1}{2} \langle \dot{\mathbf{x}} | \mathbf{M} | \dot{\mathbf{x}} \rangle \qquad = \frac{1}{2} \langle \mathbf{x} | \mathbf{K} | \mathbf{x} \rangle \quad \text{where: } \mathbf{K} = \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix}$$

equations

$$\begin{array}{ccc} u_1 & 0 \\ 0 & m_2 \end{array} \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
  
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## 2D harmonic oscillator equation solutions

1. May rewrite equation  $\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$  in acceleration matrix form:  $|\ddot{\mathbf{x}}\rangle = -\mathbf{A}|\mathbf{x}\rangle$  where:  $\mathbf{A} = \mathbf{M}^{-1} \cdot \mathbf{K}$ 

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} m_1 & 0 \\ 0 & m_2 \end{pmatrix}^{-1} \begin{pmatrix} k_1 + k_{12} & -k_{12} \\ -k_{12} & k_2 + k_{12} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{-k_{12}}{m_2} & \frac{k_2 + k_{12}}{m_2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = - \begin{pmatrix} \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \\ \frac{k_1 + k_{12}}{m_1} & \frac{-k_{12}}{m_1} \end{pmatrix}$$

2. Need to find *eigenvectors*  $|\mathbf{e}_1\rangle$ ,  $|\mathbf{e}_2\rangle$ ,... of acceleration matrix such that:  $\mathbf{A}|\mathbf{e}_n\rangle = \varepsilon_n|\mathbf{e}_n\rangle = \omega_n^2|\mathbf{e}_n\rangle$ Then equations decouple to:  $|\ddot{\mathbf{e}}_n\rangle = -\mathbf{A}|\mathbf{e}_n\rangle = -\varepsilon_n|\mathbf{e}_n\rangle = -\omega_n^2|\mathbf{e}_n\rangle$  where  $\varepsilon_n$  is an *eigenvalue* and  $\omega_n$  is an *eigenfrequency* 

# 2D harmonic oscillator equation solutions

1. May rewrite equation  $\mathbf{M} \cdot |\ddot{\mathbf{x}}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$  in *acceleration* matrix form:  $|\ddot{\mathbf{x}}\rangle = -\mathbf{A}|\mathbf{x}\rangle$  where:  $\mathbf{A} = \mathbf{M}^{-1} \cdot \mathbf{K}$ 

$$\begin{pmatrix} \ddot{x}_{1} \\ \ddot{x}_{2} \end{pmatrix} = - \begin{pmatrix} m_{1} & 0 \\ 0 & m_{2} \end{pmatrix}^{-1} \begin{pmatrix} k_{1} + k_{12} & -k_{12} \\ -k_{12} & k_{2} + k_{12} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} = - \begin{pmatrix} \frac{k_{1} + k_{12}}{m_{1}} & \frac{-k_{12}}{m_{1}} \\ \frac{-k_{12}}{m_{2}} & \frac{k_{2} + k_{12}}{m_{2}} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}$$

2. Need to find *eigenvectors*  $|\mathbf{e}_1\rangle$ ,  $|\mathbf{e}_2\rangle$ ,... of acceleration matrix such that:  $\mathbf{A}|\mathbf{e}_n\rangle = \varepsilon_n|\mathbf{e}_n\rangle = \omega_n^2|\mathbf{e}_n\rangle$ 

Then equations decouple to: 
$$\frac{d^2}{dt^2} |\mathbf{e}_1\rangle = |\ddot{\mathbf{e}}_1\rangle = -\mathbf{A} |\mathbf{e}_1\rangle = -\mathbf{E}_1 |\mathbf{e}_1\rangle = -\mathbf{\omega}_1^2 |\mathbf{e}_1\rangle \text{ so: } |\mathbf{e}_1(t)\rangle = e^{-i\omega_1 t} |\mathbf{e}_1(0)\rangle$$

where  $\varepsilon_1$  is 1<sup>st</sup> eigenvalue and  $\omega_1$  is 1<sup>st</sup> eigenfrequency

and: 
$$\frac{d^2}{dt^2} |\mathbf{e}_2\rangle \equiv |\ddot{\mathbf{e}}_2\rangle = -\mathbf{A} |\mathbf{e}_2\rangle = -\varepsilon_2 |\mathbf{e}_2\rangle = -\omega_2^2 |\mathbf{e}_2\rangle \text{ so: } |\mathbf{e}_2(t)\rangle = e^{-i\omega_2 t} |\mathbf{e}_2(0)\rangle$$
  
where  $\varepsilon_2$  is  $2^{nd}$  eigenvalue and  $\omega_2$  is  $2^{nd}$  eigenfrequency

To introduce eigensolutions we take a simple case of unit masses  $(m_1=1=m_2)$ 

So equation of motion is simply:  $|\ddot{\mathbf{x}}\rangle = -\mathbf{K}|\mathbf{x}\rangle$ 

Eigenvectors  $|\mathbf{x}\rangle = |\mathbf{e}_n\rangle$  are in special directions where  $|\ddot{\mathbf{x}}\rangle = -\mathbf{K}|\mathbf{x}\rangle$  is in same direction as  $|\mathbf{x}\rangle$ 

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*Fig. 3.3.4 Plot of potential function*  $V(x_1, x_2)$  *showing elliptical*  $V(x_1, x_2)$ *=const. level curves.* 



*Fig. 3.3.4 Plot of potential function*  $V(x_1, x_2)$  *showing elliptical*  $V(x_1, x_2)$ *=const. level curves.* 



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Spring set down for STRONGER coupling

Spring set *up* for weaker coupling





BoxIt Web Simulation - Coupled Oscillators Beating





BoxIt Web Simulation - Coupled Oscillators w/Rationals

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2D-HO beats and mixed mode geometry



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 Eigensolutions by matrix-algebra with example M= ( <sup>4</sup> 1 / <sup>3</sup> 2 ) Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues ⇒ eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P)
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An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.  $\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ 

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \cdots, |\varepsilon_n\rangle\}$  called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$ 

An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M}|\boldsymbol{\varepsilon}_{k}\rangle = \boldsymbol{\varepsilon}_{k}|\boldsymbol{\varepsilon}_{k}\rangle, \text{ or: } (\mathbf{M}-\boldsymbol{\varepsilon}_{k}\mathbf{1})|\boldsymbol{\varepsilon}_{k}\rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots |\varepsilon_n\rangle\}$  called *diagonalization* gives

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An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$  called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$ 

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left( \boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$

$$\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2 - \varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4 - \varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M})$$

An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M}|\boldsymbol{\varepsilon}_{k}\rangle = \boldsymbol{\varepsilon}_{k}|\boldsymbol{\varepsilon}_{k}\rangle, \text{ or: } (\mathbf{M}-\boldsymbol{\varepsilon}_{k}\mathbf{1})|\boldsymbol{\varepsilon}_{k}\rangle = \mathbf{0}$$

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$$\mathbf{M}|\varepsilon\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \varepsilon \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

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$$0 = \det |\mathbf{M} - \varepsilon \cdot \mathbf{l}| = \det \begin{vmatrix} 4 & 1 \\ 3 & 2 \end{vmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = \det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$
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$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let:  $\varepsilon_1 = 1$  and:  $\varepsilon_2 = 5$ 

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms



2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag) Geometry of phase and polarization

Eigensolutions by matrix-algebra with example  $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation (d projectors)Idempotent projectors (how eigenvalues  $\Rightarrow$  eigenvectors) Operator orthonormality and Completeness (Idempotent means:  $P \cdot P = P$ ) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a'-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$  called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$ 

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left( \boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$
  
Secular equation has *n*-factors, one for each eigenvalue.

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_1) (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_2) \cdots (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_n)$$

Each  $\varepsilon$  replaced by **M** and each  $\varepsilon_k$  by  $\varepsilon_k \mathbf{1}$  gives *Hamilton-Cayley* matrix equation.

$$\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1}) (\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$$

Obviously true if **M** has diagonal form. (But, that's circular logic. Faith needed!)

$$\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon}\begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

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$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1 \cdot \mathbf{1})(\mathbf{M} - 5 \cdot \mathbf{1})$$
$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6 \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms



2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag) Geometry of phase and polarization

*Eigensolutions by matrix-algebra with example*  $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ *Secular equation* 

→ Hamilton-Cayley equation and projectors
 ↓ Idempotent projectors (how eigenvalues⇒eigenvectors)
 ○ Operator orthonormality and Completeness (Idempotent means: P·P=P)
 Spectral Decompositions
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 ○ Orthonormality vs. Completeness vis-a`-vis Operator vs. State

Lagrange functional interpolation formula

Diagonalizing Transformations (D-Ttran) from projectors

An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$  called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$ 

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left( \boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$
  
Secular equation has *n*-factors, one for each eigenvalue.

det 
$$|\mathbf{M} - \varepsilon \mathbf{1}| = 0 = (-1)^n (\varepsilon - \varepsilon_1) (\varepsilon - \varepsilon_2) \cdots (\varepsilon - \varepsilon_n)$$

Each  $\varepsilon$  replaced by **M** and each  $\varepsilon_k$  by  $\varepsilon_k \mathbf{1}$  gives *Hamilton-Cayley* matrix equation.

 $\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1})(\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$ 

Obviously true if **M** has diagonal form. (But, that's circular logic. Faith needed!)

Replace *j*<sup>th</sup> HC-factor by (1) to make *projection operators* 

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - \varepsilon_{2}\mathbf{1})\cdots(\mathbf{M} - \varepsilon_{n}\mathbf{1})$$
$$\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1}\mathbf{1})(\mathbf{1})\cdots(\mathbf{M} - \varepsilon_{n}\mathbf{1})$$
$$\vdots$$
$$\mathbf{p}_{n} = (\mathbf{M} - \varepsilon_{1}\mathbf{1})(\mathbf{M} - \varepsilon_{2}\mathbf{1})\cdots(\mathbf{1})$$

$$\mathbf{M}|\varepsilon\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \varepsilon \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2-\varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4-\varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
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 so let:  $\varepsilon_1 = 1$  and:  $\varepsilon_2 = 5$ 

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1})$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1}) = \begin{pmatrix} 4-5 & 1 \\ 3 & 2-5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

$$\mathbf{p}_{2} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{1}) = \begin{pmatrix} 4-1 & 1 \\ 3 & 2-1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix}$$

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First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left( \boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

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$$|\mathbf{M} - \varepsilon \mathbf{1}| = 0 = (-1)^n (\varepsilon - \varepsilon_1) (\varepsilon - \varepsilon_2) \cdots (\varepsilon - \varepsilon_n)$$

Each  $\varepsilon$  replaced by **M** and each  $\varepsilon_k$  by  $\varepsilon_k \mathbf{1}$  gives *Hamilton-Cayley* matrix equation.

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Obviously true if M has diagonal form. (But, that's circular logic. Faith needed!)

Replace  $j^{\text{th}}$  HC-factor by (1) to make projection operators  $\mathbf{p}_{k} = \prod_{j \neq k} (\mathbf{M} - \varepsilon_{j} \mathbf{1})$   $\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - \varepsilon_{2} \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$   $\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1} \mathbf{1})(-\mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$   $\vdots$   $\mathbf{p}_{n} = (\mathbf{M} - \varepsilon_{1} \mathbf{1})(\mathbf{M} - \varepsilon_{2} \mathbf{1}) \cdots (-\mathbf{1})$ (Assume distinct e-values here: Non-degeneracy elause)  $\varepsilon_{j} \neq \varepsilon_{k} \neq \dots$ 

Each  $\mathbf{p}_k$  contains *eigen-bra-kets* since:  $(\mathbf{M} - \varepsilon_k \mathbf{1})\mathbf{p}_k = 0$  or:  $\mathbf{M}\mathbf{p}_k = \varepsilon_k \mathbf{p}_k = \mathbf{p}_k \mathbf{M}$ .

 $\mathbf{M}|\boldsymbol{\varepsilon}\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \boldsymbol{\varepsilon} \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\boldsymbol{\varepsilon} & 1 \\ 3 & 2-\boldsymbol{\varepsilon} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ 

Trying to solve by Kramer's inversion:

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$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
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$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1}) = \begin{pmatrix} 4-5 & 1 \\ 3 & 2-5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

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$$\mathbf{Mp}_{1} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1\cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1\cdot \mathbf{p}_{1}$$

$$\mathbf{Mp}_{2} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5\cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5\cdot \mathbf{p}_{2}$$

An *eigenvector*  $|\varepsilon_k\rangle$  of **M** is in a direction that is left unchanged by **M**.

$$\mathbf{M} | \boldsymbol{\varepsilon}_{k} \rangle = \boldsymbol{\varepsilon}_{k} | \boldsymbol{\varepsilon}_{k} \rangle, \text{ or: } (\mathbf{M} - \boldsymbol{\varepsilon}_{k} \mathbf{1}) | \boldsymbol{\varepsilon}_{k} \rangle = \mathbf{0}$$

 $\varepsilon_k$  is *eigenvalue* associated with eigenvector  $|\varepsilon_k\rangle$  direction. A change of basis to  $\{|\varepsilon_1\rangle, |\varepsilon_2\rangle, \dots, |\varepsilon_n\rangle\}$  called *diagonalization* gives

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{2} \rangle & \cdots & \langle \boldsymbol{\varepsilon}_{n} | \mathbf{M} | \boldsymbol{\varepsilon}_{n} \rangle \end{pmatrix} = \begin{pmatrix} \boldsymbol{\varepsilon}_{1} & 0 & \cdots & 0 \\ 0 & \boldsymbol{\varepsilon}_{2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \boldsymbol{\varepsilon}_{n} \end{pmatrix}$ 

First step in finding eigenvalues: Solve secular equation

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n \left( \boldsymbol{\varepsilon}^n + a_1 \boldsymbol{\varepsilon}^{n-1} + a_2 \boldsymbol{\varepsilon}^{n-2} + \dots + a_{n-1} \boldsymbol{\varepsilon} + a_n \right)$$

where:

$$a_1 = -Trace \mathbf{M}, \dots, a_k = (-1)^k \sum \text{diagonal k-by-k minors of } \mathbf{M}, \dots, a_n = (-1)^n \det |\mathbf{M}|$$
  
Secular equation has *n*-factors, one for each eigenvalue.

$$\det |\mathbf{M} - \boldsymbol{\varepsilon} \mathbf{1}| = 0 = (-1)^n (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_1) (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_2) \cdots (\boldsymbol{\varepsilon} - \boldsymbol{\varepsilon}_n)$$

Each  $\varepsilon$  replaced by **M** and each  $\varepsilon_k$  by  $\varepsilon_k \mathbf{1}$  gives *Hamilton-Cayley* matrix equation.  $\mathbf{0} = (\mathbf{M} - \varepsilon_1 \mathbf{1})(\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_n \mathbf{1})$ 

Obviously true if M has diagonal form. (But, that's circular logic. Faith needed!)

Replace  $j^{\text{th}}$  HC-factor by (1) to make *projection operators*  $\mathbf{p}_{k} = \prod_{j \neq k} (\mathbf{M} - \varepsilon_{j} \mathbf{1})$   $\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - \varepsilon_{2} \mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$   $\mathbf{p}_{2} = (\mathbf{M} - \varepsilon_{1} \mathbf{1})(-\mathbf{1}) \cdots (\mathbf{M} - \varepsilon_{n} \mathbf{1})$  (Assume <u>distinct</u> e-values here: *Non-degeneracy clause*)  $\varepsilon_{j} \neq \varepsilon_{k} \neq \dots$ 

 $\mathbf{p}_n = (\mathbf{M} - \varepsilon_1 \mathbf{1}) (\mathbf{M} - \varepsilon_2 \mathbf{1}) \cdots (\mathbf{1})$ 

Each  $\mathbf{p}_k$  contains *eigen-bra-kets* since:  $(\mathbf{M} - \varepsilon_k \mathbf{1})\mathbf{p}_k = 0$  or:  $\mathbf{M}\mathbf{p}_k = \varepsilon_k \mathbf{p}_k = \mathbf{p}_k \mathbf{M}$ .

since  $\mathbf{M}^{1}$ ,  $\mathbf{M}^{2}$ ,..commute with  $\mathbf{M}$ .

*Notice*  $\mathbf{p}_k$  *commutes with*  $\mathbf{M}_{,...}$ 

$$\mathbf{M}|\varepsilon\rangle = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \varepsilon \begin{pmatrix} x \\ y \end{pmatrix} \text{ or: } \begin{pmatrix} 4-\varepsilon & 1 \\ 3 & 2-\varepsilon \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Trying to solve by Kramer's inversion:

$$x = \frac{\det \begin{vmatrix} 0 & 1 \\ 0 & 2 - \varepsilon \end{vmatrix}}{\det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}} \quad \text{and} \quad y = \frac{\det \begin{vmatrix} 4 - \varepsilon & 0 \\ 3 & 0 \end{vmatrix}}{\det \begin{vmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{vmatrix}}$$

$$0 = \det \left| \mathbf{M} - \varepsilon \cdot \mathbf{1} \right| = \det \left| \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} - \varepsilon \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right| = \det \left| \begin{pmatrix} 4 - \varepsilon & 1 \\ 3 & 2 - \varepsilon \end{pmatrix} \right|$$
$$0 = (4 - \varepsilon)(2 - \varepsilon) - 1 \cdot 3 = 8 - 6\varepsilon + \varepsilon^2 - 1 \cdot 3 = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = \varepsilon^2 - Trace(\mathbf{M})\varepsilon + \det(\mathbf{M}) = \varepsilon^2 - 6\varepsilon + 5$$

$$0 = (\varepsilon - 1)(\varepsilon - 5)$$
 so let:  $\varepsilon_1 = 1$  and:  $\varepsilon_2 = 5$ 

$$0 = \mathbf{M}^{2} - 6\mathbf{M} + 5\mathbf{1} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1})$$

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}^{2} - 6\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} + 5\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\mathbf{p}_{1} = (\mathbf{1})(\mathbf{M} - 5\cdot\mathbf{1}) = \begin{pmatrix} 4-5 & 1 \\ 3 & 2-5 \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix}$$

$$\mathbf{p}_{2} = (\mathbf{M} - 1\cdot\mathbf{1})(\mathbf{1}) = \begin{pmatrix} 4-1 & 1 \\ 3 & 2-1 \end{pmatrix} = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix}$$

$$\mathbf{Mp}_{1} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1\cdot \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} = 1\cdot \mathbf{p}_{1}$$

$$\mathbf{Mp}_{2} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5\cdot \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = 5\cdot \mathbf{p}_{2}$$

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms



2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag) Geometry of phase and polarization

Eigensolutions by matrix-algebra with example M = ( <sup>4</sup> 1 <sub>3</sub> <sup>2</sup> ) Secular equation Hamilton-Cayley equation and projectors
→ Idempotent projectors ( <del>An eigenvalues → eigenvectors)</del>. Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{j} \mathbf{p}_{k} = \mathbf{p}_{j} \prod_{m \neq k} (\mathbf{M} - \varepsilon_{m} \mathbf{1}) = \prod_{m \neq k} (\mathbf{p}_{j} \mathbf{M} - \varepsilon_{m} \mathbf{p}_{j} \mathbf{1}) & \mathbf{M} \mathbf{p}_{k} = \varepsilon_{k} \mathbf{p}_{k} = \mathbf{p}_{k} \mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{j} : \\ \mathbf{p}_{j} \mathbf{p}_{k} = \prod_{m \neq k} (\varepsilon_{j} \mathbf{p}_{j} - \varepsilon_{m} \mathbf{p}_{j}) = \mathbf{p}_{j} \prod_{m \neq k} (\varepsilon_{j} - \varepsilon_{m}) = \begin{cases} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{k} \prod_{m \neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} & \mathbf{P}_{k} = k \end{aligned}$$

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}} & \text{With example matrix} & \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k} (\mathbf{M} - \varepsilon_{m}\mathbf{1}) = \prod_{m\neq k} (\mathbf{p}_{j}\mathbf{M} - \varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ \text{Multiplication properties of } \mathbf{p}_{j}: \\ \mathbf{p}_{j}\mathbf{p}_{k} = \prod_{m\neq k} (\varepsilon_{j}\mathbf{p}_{j} - \varepsilon_{m}\mathbf{p}_{j}) = \mathbf{p}_{j}\prod_{m\neq k} (\varepsilon_{j} - \varepsilon_{m}) = \begin{cases} \mathbf{0} & \text{if } : j \neq k \\ \mathbf{p}_{k}\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} \\ \mathbf{p}_{k}\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m}) & \text{if } : j = k \end{cases} \\ \text{Last step:} \\ \text{make Idempotent Projectors: } \mathbf{P}_{k} = \frac{\mathbf{p}_{k}}{\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m})} = \frac{\prod_{m\neq k} (\mathbf{M} - \varepsilon_{m}\mathbf{1})}{\prod_{m\neq k} (\varepsilon_{k} - \varepsilon_{m})} \\ \mathbf{P}_{1} = \frac{(\mathbf{M} - \mathbf{5} \cdot \mathbf{1})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \end{aligned}$$

$$\begin{array}{ll} \text{Matrix-algebraic method for finding eigenvector and eigenvalues} \\ \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k}(\mathbf{M}-\varepsilon_{m}\mathbf{1}) = \prod_{m\neq k}(\mathbf{p}_{j}\mathbf{M}-\varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) \\ \text{Multiplication properties of } \mathbf{p}_{j}: \\ \mathbf{p}_{j}\mathbf{p}_{k} = \prod_{m\neq k}(\varepsilon_{j}\mathbf{p}_{j}-\varepsilon_{m}\mathbf{p}_{j}) = \mathbf{p}_{j}\prod_{m\neq k}(\varepsilon_{j}-\varepsilon_{m}) = \begin{cases} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k}\prod_{m\neq k}(\varepsilon_{k}-\varepsilon_{m}) & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k}\prod_{m\neq k}(\varepsilon_{k}-\varepsilon_{m}) & \text{if } : j=k \end{cases} \\ \mathbf{p}_{i}\mathbf{p}_{k} = \begin{bmatrix} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k}\prod_{m\neq k}(\varepsilon_{k}-\varepsilon_{m}) & \text{if } : j=k \end{cases} \\ \mathbf{p}_{i}\mathbf{p}_{k} = \begin{cases} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k}\prod_{m\neq k}(\varepsilon_{k}-\varepsilon_{m}) & \text{if } : j=k \end{cases} \\ \mathbf{p}_{i}\mathbf{p}_{k} = \begin{bmatrix} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \begin{bmatrix} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \begin{bmatrix} \mathbf{0} & \text{if } : j\neq k \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} & \text{if } : j=k \end{cases} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} \\ \mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} \\ \mathbf{p}_{k$$

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms



2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag) Geometry of phase and polarization

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$$\begin{aligned} & \text{Matrix-algebraic method for finding eigenvector and eigenvalues} \\ & \mathbf{p}_{j}\mathbf{p}_{k} = \mathbf{p}_{j}\prod_{m\neq k} (\mathbf{M}-\varepsilon_{m}\mathbf{1}) = \prod_{m\neq k} (\mathbf{p}_{j}\mathbf{M}-\varepsilon_{m}\mathbf{p}_{j}\mathbf{1}) \\ & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ & \mathbf{M}\mathbf{p}_{k} = \varepsilon_{k}\mathbf{p}_{k} = \mathbf{p}_{k}\mathbf{M} \\ & \mathbf{p}_{1} = (\mathbf{M}-5\cdot\mathbf{1}) = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} \\ & \mathbf{p}_{1} = (\mathbf{M}-5\cdot\mathbf{1}) = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ 3 & 1 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ & \mathbf{p}_{2} = (\mathbf{M}-1\cdot\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ & -\frac{3}{2} \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ & -\frac{3}{2} \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ & -\frac{3}{2} \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ & -\frac{3}{2} \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2} = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ & -\frac{3}{2} \end{pmatrix} \\ & \mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{2}\mathbf{p}_{1}\mathbf{p}_{2}\mathbf{p}_{$$

$$\begin{aligned} \text{Matrix-algebraic method for finding eigenvector and eigenvalues}}_{\mathbf{p}_{1}\mathbf{p}_{x}} & \text{With example matrix} \quad \mathbf{M} = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \\ \mathbf{p}_{1}\mathbf{p}_{x} = \mathbf{p}_{1} \prod_{n=1}^{\infty} (\mathbf{p}_{1}\mathbf{M} - \varepsilon_{n}\mathbf{p}_{1}) = \prod_{n=1}^{\infty} (\mathbf{p}_{1}\mathbf{M} - \varepsilon_{n}\mathbf{p}_{1}) \\ \mathbf{M}\mathbf{p}_{1} = \varepsilon_{k}\mathbf{p}_{1} = \mathbf{p}_{k} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} -1 & 1 \\ 3 & -3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 3 & 1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 4 & -1 \\ 3 & -3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{1}) = \begin{pmatrix} 4 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{1}) = \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{H}) = \begin{pmatrix} 5 & -1 \\ -3 & 3 \end{pmatrix} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{2} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{2} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{1} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{2} = (\mathbf{M} - 1\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{1} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{2} = (\mathbf{M} - 5\mathbf{H}) = \mathbf{p}_{2} \\ \mathbf{p}_{2} = (\mathbf{p}_{2} - \mathbf{p}_{2} \\$$

2D harmonic oscillator (2D-HO) equations of motion Lagrangian and matrix forms



2D harmonic oscillator equation eigensolutions (normal modes) Eigensolutions by geometry for 2D-HO with bilateral (B-Type) symmetry Symmetric (low frequency) mode versus antisymmetric (high frequency) mode Mixed mode beat dynamics (with constant  $\pi/2$  phase-lag) Geometry of phase and polarization

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2D harmonic oscillator equation eigensolutions Geometric method
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2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed  $\pi/2$  phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

ANALOGY: 2-State Schrodinger:  $i\hbar \partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  versus Classical 2D-HO:  $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ Hamilton-Pauli spinor symmetry (ABCD-Types)





2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M=  $\begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒ eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) ✓ Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors 2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed π/2 phase

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2D harmonic oscillator equation eigensolutions Geometric method
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*ANALOGY: 2-State Schrodinger:*  $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  versus *Classical 2D-HO:*  $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)* 







2D harmonic oscillator equation eigensolutions *Geometric method Matrix-algebraic eigensolutions with example*  $M = \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix}$ Secular equation Hamilton-Cayley equation and projectors *Idempotent projectors (how eigenvalues*  $\Rightarrow$  *eigenvectors) Operator orthonormality and Completeness (Idempotent means:* **P**·**P**=**P**) Spectral Decompositions Functional spectral decomposition - Orthonormality vs. Completeness vis-a'-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors 2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed  $\pi/2$  phase 2D-HO eigensolution example with asymmetric (A-Type) symmetry

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Initial state projection, mixed mode beat dynamics with variable phase



Thursday, March 10, 2016

Orthonormality vs. Completeness vis-a`-vis Operator vs. State Operator expressions for orthonormality appear quite different from expressions for completeness.  $\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \ldots + \mathbf{P}_{n}$  Orthonormality vs. Completeness vis-a`-vis Operator vs. State Operator expressions for orthonormality appear quite different from expressions for completeness.  $\mathbf{P}_{j}\mathbf{P}_{k} = \delta_{jk}\mathbf{P}_{k} = \begin{cases} \mathbf{0} & if: j \neq k \\ \mathbf{P}_{k} & if: j = k \end{cases} \qquad \mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n}$ 

 $|\varepsilon_{j}\rangle\langle\varepsilon_{j}|\varepsilon_{k}\rangle\langle\varepsilon_{k}|=\delta_{jk}|\varepsilon_{k}\rangle\langle\varepsilon_{k}| \text{ or: } \langle\varepsilon_{j}|\varepsilon_{k}\rangle=\delta_{jk} \qquad \mathbf{1}=|\varepsilon_{1}\rangle\langle\varepsilon_{1}|+|\varepsilon_{2}\rangle\langle\varepsilon_{2}|+...+|\varepsilon_{n}\rangle\langle\varepsilon_{n}|$ 

Orthonormality vs. Completeness vis-a`-vis Operator vs. State

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State vector representations of orthonormality are quite **similar** to representations of completeness. *Like 2-sides of the same coin.* 

 $\{|x\rangle, |y\rangle\} \text{-orthonormality with } \{|\varepsilon_1\rangle, |\varepsilon_2\rangle\} \text{-completeness}$  $\langle x|y\rangle = \delta_{x,y} = \langle x|\mathbf{1}|y\rangle = \langle x|\varepsilon_1\rangle \langle \varepsilon_1|y\rangle + \langle x|\varepsilon_2\rangle \langle \varepsilon_2|y\rangle.$ 

 $\{|\varepsilon_{I}\rangle, |\varepsilon_{2}\rangle\} \text{-orthonormality with } \{|x\rangle, |y\rangle\} \text{-completeness}$  $\langle \varepsilon_{i}|\varepsilon_{j}\rangle = \delta_{i,j} = \langle \varepsilon_{i}|\mathbf{1}|\varepsilon_{j}\rangle = \langle \varepsilon_{i}|x\rangle\langle x|\varepsilon_{j}\rangle + \langle \varepsilon_{i}|y\rangle\langle y|\varepsilon_{j}\rangle$
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However Schrodinger wavefunction notation  $\psi(x) = \langle x | \psi \rangle$  shows quite a difference...

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However Schrodinger wavefunction notation  $\psi(x) = \langle x | \psi \rangle$  shows quite a difference... ...particularly in the orthonormality integral. 2D harmonic oscillator equations Lagrangian and matrix forms and Reciprocity symmetry

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$$\mathbf{1} = \mathbf{P}_{1} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

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A Proof of Projector Completeness (Truer-than-true) Compare matrix completeness relation and functional spectral decompositions

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If f(x) happens to be a polynomial of degree N-1 or less, then L(f(x)) = f(x) may be exact everywhere.

$$1 = \sum_{m=1}^{N} P_m(x) \qquad x = \sum_{m=1}^{N} x_m P_m(x) \qquad x^2 = \sum_{m=1}^{N} x_m^2 P_m(x)$$

Compare matrix *completeness relation* and *functional spectral decompositions*  $\Pi(x, y)$ 

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One point determines a constant level line,



Compare matrix completeness relation and functional spectral decompositions

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One point determines a constant level line, two separate points uniquely determine a sloping line,



Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

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One point determines a constant level line, two separate points uniquely determine a sloping line,

three separate points uniquely determine a parabola, etc.



Compare matrix completeness relation and functional spectral decompositions

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*Lagrange interpolation formula*  $\rightarrow$  *Completeness formula* as  $x \rightarrow M$  and as  $x_k \rightarrow \varepsilon_k$  and as  $P_k(x_k) \rightarrow P_k$ 

Compare matrix completeness relation and functional spectral decompositions

$$\mathbf{1} = \mathbf{P}_{l} + \mathbf{P}_{2} + \dots + \mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} \mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} \frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})} \qquad f(\mathbf{M}) = f(\boldsymbol{\varepsilon}_{1})\mathbf{P}_{1} + f(\boldsymbol{\varepsilon}_{2})\mathbf{P}_{2} + \dots + f(\boldsymbol{\varepsilon}_{n})\mathbf{P}_{n} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\mathbf{P}_{k} = \sum_{\boldsymbol{\varepsilon}_{k}} f(\boldsymbol{\varepsilon}_{k})\frac{\prod_{m \neq k} (\mathbf{M} - \boldsymbol{\varepsilon}_{m} \mathbf{1})}{\prod_{m \neq k} (\boldsymbol{\varepsilon}_{k} - \boldsymbol{\varepsilon}_{m})}$$

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$$\mathbf{P}_{1} + \mathbf{P}_{2} = \frac{\prod_{j \neq 1} \left(\mathbf{M} - \varepsilon_{j}\mathbf{1}\right)}{\prod_{j \neq 1} \left(\varepsilon_{1} - \varepsilon_{j}\right)} + \frac{\prod_{j \neq 1} \left(\mathbf{M} - \varepsilon_{j}\mathbf{1}\right)}{\prod_{j \neq 1} \left(\varepsilon_{2} - \varepsilon_{j}\right)} = \frac{\left(\mathbf{M} - \varepsilon_{2}\mathbf{1}\right)}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} + \frac{\left(\mathbf{M} - \varepsilon_{1}\mathbf{1}\right)}{\left(\varepsilon_{2} - \varepsilon_{1}\right)} = \frac{\left(\mathbf{M} - \varepsilon_{2}\mathbf{1}\right) - \left(\mathbf{M} - \varepsilon_{1}\mathbf{1}\right)}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} = \frac{-\varepsilon_{2}\mathbf{1} + \varepsilon_{1}\mathbf{1}}{\left(\varepsilon_{1} - \varepsilon_{2}\right)} = \mathbf{1} \text{ (for all } \varepsilon_{j})$$

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However, only *select* values  $\varepsilon_k$  work for eigen-forms  $\mathbf{MP}_k = \varepsilon_k \mathbf{P}_k$  or orthonormality  $\mathbf{P}_j \mathbf{P}_k = \delta_{jk} \mathbf{P}_k$ .

2D harmonic oscillator equations Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method
Matrix-algebraic eigensolutions with example M=(4 1) Secular equation (3 2)
Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors)
Operator orthonormality and Completeness (Idempotent means: P·P=P)
Spectral Decompositions Functional spectral decomposition
Orthonormality vs. Completeness vis-a`-vis Operator vs. State Lagrange functional interpolation formula
Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed  $\pi/2$  phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

*ANALOGY: 2-State Schrodinger:*  $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  versus *Classical 2D-HO:*  $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry* (*ABCD-Types*) Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors.  $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{l})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\left(\frac{1}{2} & -\frac{1}{2}\right)}{k_{1}} = |\varepsilon_{1}\rangle\langle\varepsilon_{1}|$   $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{l})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\left(\frac{3}{2} - \frac{1}{2}\right)}{k_{2}} = |\varepsilon_{2}\rangle\langle\varepsilon_{2}|$  Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors.  $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{l})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\left(\frac{1}{2} & -\frac{1}{2}\right)}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$   $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{l})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\left(\frac{3}{2} & \frac{1}{2}\right)}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$ 

Load distinct bras  $\langle \varepsilon_1 |$  and  $\langle \varepsilon_2 |$  into d-tran rows, kets  $|\varepsilon_1 \rangle$  and  $|\varepsilon_2 \rangle$  into inverse d-tran columns.

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$$\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left( \begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left( \begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\} , \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left( \begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right), \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left( \begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right) \right\} \right\}$$

 $\begin{array}{ll} (\boldsymbol{\varepsilon}_{1},\boldsymbol{\varepsilon}_{2}) \leftarrow (1,2) \ d\text{-Tran matrix} & (1,2) \leftarrow (\boldsymbol{\varepsilon}_{1},\boldsymbol{\varepsilon}_{2}) \ \text{INVERSE } d\text{-Tran matrix} \\ \left( \begin{array}{c} \left\langle \boldsymbol{\varepsilon}_{1} \middle| x \right\rangle & \left\langle \boldsymbol{\varepsilon}_{1} \middle| y \right\rangle \\ \left\langle \boldsymbol{\varepsilon}_{2} \middle| x \right\rangle & \left\langle \boldsymbol{\varepsilon}_{2} \middle| y \right\rangle \end{array} \right) = \left( \begin{array}{c} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{array} \right) & , \\ \left( \begin{array}{c} \left\langle x \middle| \boldsymbol{\varepsilon}_{1} \right\rangle & \left\langle x \middle| \boldsymbol{\varepsilon}_{2} \right\rangle \\ \left\langle y \middle| \boldsymbol{\varepsilon}_{1} \right\rangle & \left\langle y \middle| \boldsymbol{\varepsilon}_{2} \right\rangle \end{array} \right) = \left( \begin{array}{c} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{array} \right) \end{array}$ 

 $\begin{array}{c} (\boldsymbol{\varepsilon}_{1},\boldsymbol{\varepsilon}_{2}) \leftarrow (1,2) \ d\text{-Tran matrix} \\ \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | x \rangle & \langle \boldsymbol{\varepsilon}_{1} | y \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | x \rangle & \langle \boldsymbol{\varepsilon}_{2} | y \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle x | \boldsymbol{\varepsilon}_{1} \rangle & \langle x | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle y | \boldsymbol{\varepsilon}_{1} \rangle & \langle y | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} \\ \text{Use Dirac labeling for all components so transformation is OK} \\ \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | x \rangle & \langle \boldsymbol{\varepsilon}_{1} | y \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | x \rangle & \langle \boldsymbol{\varepsilon}_{2} | y \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle x | \mathbf{K} | x \rangle & \langle x | \mathbf{K} | y \rangle \\ \langle y | \mathbf{K} | x \rangle & \langle y | \mathbf{K} | y \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle x | \boldsymbol{\varepsilon}_{1} \rangle & \langle x | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle y | \boldsymbol{\varepsilon}_{1} \rangle & \langle y | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} \\ \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} & \cdot \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} & \cdot \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 5 \end{pmatrix} \end{array}$ 

Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors.  $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{I})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \end{pmatrix}}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$  $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{3}{2} & \frac{1}{2} \end{pmatrix}}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle \langle \boldsymbol{\varepsilon}_{2}|$ Load distinct bras  $\langle \varepsilon_1 |$  and  $\langle \varepsilon_2 |$  into d-tran rows, kets  $|\varepsilon_1 \rangle$  and  $|\varepsilon_2 \rangle$  into <u>inverse</u> d-tran columns.  $\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left( \begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left( \begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\}, \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right|, \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right| \right\}$  $(\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2) \leftarrow (1, 2) d$ -Tran matrix  $(1,2) \leftarrow (\varepsilon_1, \varepsilon_2)$  INVERSE *d*-Tran matrix  $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_2 \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_2 \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix}$ Use Dirac labeling for all components so transformation is OK  $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}$  $\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} \qquad \cdot \qquad \begin{pmatrix} 4 & 1 \\ 3 & 2 \end{pmatrix} \qquad \cdot \qquad \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} \qquad = \qquad \begin{pmatrix} 1 & 0 \\ 0 & 5 \end{pmatrix}$ Check inverse-d-tran is really inverse of your d-tran.

 $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | 1 \rangle & \langle \boldsymbol{\varepsilon}_{1} | 2 \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | 1 \rangle & \langle \boldsymbol{\varepsilon}_{2} | 2 \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle 1 | \boldsymbol{\varepsilon}_{1} \rangle & \langle 1 | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle 2 | \boldsymbol{\varepsilon}_{1} \rangle & \langle 2 | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | 1 | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | 1 | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | 1 | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | 1 | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}$  $\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} \cdot \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ 

Diagonalizing Transformations (D-Ttran) from projectors Given our eigenvectors and their Projectors.  $\mathbf{P}_{1} = \frac{(\mathbf{M} - 5 \cdot \mathbf{I})}{(1 - 5)} = \frac{1}{4} \begin{pmatrix} 1 & -1 \\ -3 & 3 \end{pmatrix} = k_{1} \begin{pmatrix} \frac{1}{2} \\ -\frac{3}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \end{pmatrix}}{k_{1}} = |\boldsymbol{\varepsilon}_{1}\rangle\langle\boldsymbol{\varepsilon}_{1}|$  $\mathbf{P}_{2} = \frac{(\mathbf{M} - 1 \cdot \mathbf{1})}{(5 - 1)} = \frac{1}{4} \begin{pmatrix} 3 & 1 \\ 3 & 1 \end{pmatrix} = k_{2} \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \otimes \frac{\begin{pmatrix} \frac{3}{2} & \frac{1}{2} \end{pmatrix}}{k_{2}} = |\boldsymbol{\varepsilon}_{2}\rangle\langle\boldsymbol{\varepsilon}_{2}|$ Load distinct bras  $\langle \varepsilon_1 |$  and  $\langle \varepsilon_2 |$  into d-tran rows, kets  $|\varepsilon_1 \rangle$  and  $|\varepsilon_2 \rangle$  into <u>inverse</u> d-tran columns.  $\left\{ \left\langle \boldsymbol{\varepsilon}_{1} \right| = \left( \begin{array}{cc} \frac{1}{2} & -\frac{1}{2} \end{array} \right), \left\langle \boldsymbol{\varepsilon}_{2} \right| = \left( \begin{array}{cc} \frac{3}{2} & \frac{1}{2} \end{array} \right) \right\}, \quad \left\{ \left| \boldsymbol{\varepsilon}_{1} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ -\frac{3}{2} \end{array} \right|, \left| \boldsymbol{\varepsilon}_{2} \right\rangle = \left| \begin{array}{cc} \frac{1}{2} \\ \frac{1}{2} \end{array} \right| \right\}$  $(\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2) \leftarrow (1, 2) d$ -Tran matrix  $(1,2) \leftarrow (\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2)$  INVERSE *d*-Tran matrix  $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_1 | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_2 | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{pmatrix} , \quad \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_2 \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_1 \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_2 \rangle \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{pmatrix}$ Use Dirac labeling for all components so transformation is OK  $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{x} | \mathbf{K} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{x} \rangle & \langle \boldsymbol{y} | \mathbf{K} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \mathbf{K} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}$  $\left(\begin{array}{ccc} \frac{1}{2} & -\frac{1}{2} \\ \frac{3}{2} & \frac{1}{2} \end{array}\right) \qquad \cdot \qquad \left(\begin{array}{ccc} 4 & 1 \\ 3 & 2 \end{array}\right) \qquad \cdot \qquad \left(\begin{array}{ccc} \frac{1}{2} & \frac{1}{2} \\ -\frac{3}{2} & \frac{1}{2} \end{array}\right) \qquad = \qquad \left(\begin{array}{ccc} 1 & 0 \\ 0 & 5 \end{array}\right)$ Check inverse-d-tran is really inverse of your d-tran. In standard quantum matrices inverses are "easy"  $\begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{y} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} \cdot \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{1} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{\varepsilon}_{1} | \boldsymbol{1} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{z}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{z} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{x} \rangle & \langle \boldsymbol{\varepsilon}_{2} | \boldsymbol{y} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle & \langle \boldsymbol{y} | \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{1} \rangle & \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{x} | \boldsymbol{\varepsilon}_{2} \rangle & \langle \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{\varepsilon} | \boldsymbol{\varepsilon}_{2} \rangle & \langle \boldsymbol{\varepsilon}_{2} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{\varepsilon} | \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon}_{2} \rangle & \langle \boldsymbol{\varepsilon}_{2} \rangle \\ \langle \boldsymbol{\varepsilon} | \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle \end{pmatrix} = \begin{pmatrix} \langle \boldsymbol{\varepsilon} | \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle \end{pmatrix}^{\dagger} = \begin{pmatrix} \langle \boldsymbol{\varepsilon} | \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle & \langle \boldsymbol{\varepsilon} \rangle \end{pmatrix} \right$  2D harmonic oscillator equations

Lagrangian and matrix forms and Reciprocity symmetry

2D harmonic oscillator equation eigensolutions Geometric method Matrix-algebraic eigensolutions with example M=(4 1 Secular equation (3 2) Hamilton-Cayley equation and projectors Idempotent projectors (how eigenvalues⇒eigenvectors) Operator orthonormality and Completeness (Idempotent means: P·P=P) Spectral Decompositions Functional spectral decomposition Orthonormality vs. Completeness vis-a'-vis Operator vs. State Lagrange functional interpolation formula Diagonalizing Transformations (D-Ttran) from projectors

2D-HO eigensolution example with bilateral (B-Type) symmetry  $\checkmark$ Mixed mode beat dynamics and fixed  $\pi/2$  phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

*ANALOGY: 2-State Schrodinger:*  $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  versus *Classical 2D-HO:*  $\partial^2_t \mathbf{X} = -\mathbf{K} \cdot \mathbf{X}$ *Hamilton-Pauli spinor symmetry (ABCD-Types)* 



Eigenvalues  $K_k$  and squared eigenfrequencies  $\omega_0(\varepsilon_k)^2$   $K_1 = \omega_0^2(\varepsilon_1) = 9$ ,  $K_2 = \omega_0^2(\varepsilon_2) = 11$ ,



$$\mathbf{P}_{1} = \frac{\begin{pmatrix} K_{11} - K_{2} & K_{12} \\ K_{12} & K_{22} - K_{2} \end{pmatrix}}{K_{1} - K_{2}} = \frac{\begin{pmatrix} 10 - 11 & -1 \\ -1 & 10 - 11 \end{pmatrix}}{9 - 11} = \frac{\begin{pmatrix} 1 & +1 \\ +1 & 1 \end{pmatrix}}{2} \qquad \qquad \mathbf{P}_{2} = \frac{\begin{pmatrix} K_{11} - K_{1} & K_{12} \\ K_{12} & K_{22} - K_{1} \end{pmatrix}}{K_{2} - K_{1}} = \frac{\begin{pmatrix} 10 - 9 & -1 \\ -1 & 10 - 9 \end{pmatrix}}{11 - 9} = \frac{\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}}{2}$$



Eigenbra vectors:  $\langle \varepsilon_1 | = (1/\sqrt{2} + 1/\sqrt{2}), \langle \varepsilon_2 | = (1/\sqrt{2} - 1/\sqrt{2})$ 



$$\begin{array}{rcl} \text{Mixed mode dynamics} \\ |x(t)\rangle &= |\varepsilon_1\rangle & \langle \varepsilon_1 | x(0) \rangle e^{-i\omega_1 t} + |\varepsilon_2\rangle & \langle \varepsilon_2 | x(0) \rangle e^{-i\omega_2 t} \\ & \left( \begin{array}{c} x_1(t) \\ x_1(t) \end{array} \right) & \left( \begin{array}{c} 1/\sqrt{2} \\ 1/\sqrt{2} \end{array} \right) \langle z_1 | z(0) \rangle & z_1 = i\omega_1 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_1 = i\omega_2 t \\ z_2 = i\omega_2 t \\ z_2$$

Eigenbra vectors:  $\langle \varepsilon_1 | = (1/\sqrt{2} + 1/\sqrt{2}), \langle \varepsilon_2 | = (1/\sqrt{2} - 1/\sqrt{2})$ 

$$\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \langle \varepsilon_1 | x(0) \rangle e^{-i\omega_1 t} + \begin{pmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \langle \varepsilon_2 | x(0) \rangle e^{-i\omega_1 t}$$



Fig. 3.3.9 Beats in weakly coupled symmetric oscillators with equal mode magnitudes.



Fig. 3.3.9 Beats in weakly coupled symmetric oscillators with equal mode magnitudes.

2D harmonic oscillator equations

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 $Det(\mathbf{K}) = 7.13 - 27 = 91 - 27 = 64$  $Trace(\mathbf{K}) = 7 + 13 = 20$ 









 $\mathbf{P}_{1} = \frac{\begin{pmatrix} K_{11} - K_{2} & K_{12} \\ K_{12} & K_{22} - K_{2} \end{pmatrix}}{K_{1} - K_{2}} = \frac{\begin{pmatrix} 7 - 16 & -3\sqrt{3} \\ -3\sqrt{3} & 13 - 16 \end{pmatrix}}{4 - 16} = \frac{\begin{pmatrix} 9 & +3\sqrt{3} \\ +3\sqrt{3} & 3 \end{pmatrix}}{12}$  $= \frac{\begin{pmatrix} 3 & \sqrt{3} \\ \sqrt{3} & 1 \end{pmatrix}}{4} = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix} (\sqrt{3}/2 & 1/2) = |\varepsilon_{1}\rangle\langle\varepsilon_{1}|$ 



$$= \frac{\begin{pmatrix} 3 & \sqrt{3} \\ \sqrt{3} & 1 \\ 4 & = \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix}} \begin{pmatrix} \sqrt{3}/2 & 1/2 \\ 1/2 \end{pmatrix} (\sqrt{3}/2 & 1/2 \end{pmatrix} = |\varepsilon_1\rangle\langle\varepsilon_1|$$

$$= \frac{\begin{pmatrix} 1 & -\sqrt{3} \\ -\sqrt{3} & 3 \\ 4 & = \begin{pmatrix} -1/2 \\ \sqrt{3}/2 \\ 4 & = \begin{pmatrix} -1/2 \\ \sqrt{3}/2 \end{pmatrix} (-1/2 & \sqrt{3}/2 \end{pmatrix} = |\varepsilon_2\rangle\langle\varepsilon_2|$$


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Thursday, March 10, 2016





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2D-HO eigensolution example with bilateral (B-Type) symmetry Mixed mode beat dynamics and fixed  $\pi/2$  phase

2D-HO eigensolution example with asymmetric (A-Type) symmetry Initial state projection, mixed mode beat dynamics with variable phase

→ ANALOGY: 2-State Schrodinger:  $i\hbar\partial_t |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  versus Classical 2D-HO:  $\partial_t^2 \mathbf{x} = -\mathbf{K} \cdot \mathbf{x}$  Hamilton-Pauli spinor symmetry (ABCD-Types)

First start with 2-by-2 Hermitian (self-conjugate) matrix

$$\mathbf{H} = \left(\begin{array}{cc} A & B - iC \\ B + iC & D \end{array}\right) = \mathbf{H}^{\dagger}$$

 $H_{jk}$  matrix must obey:  $(H_{jk})^* = H_{kj}$ 





to convert the complex 1<sup>st</sup>-order equation  $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of <u>real</u> 1<sup>st</sup>-order differential equations.







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that operates on 2-D complex Dirac ket vector  $\left|\Psi\right\rangle$  .

$$|\Psi\rangle = \begin{pmatrix} \Psi_1 \\ \Psi_2 \end{pmatrix} = \begin{pmatrix} x_1 + ip_1 \\ x_2 + ip_2 \end{pmatrix} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}$$

Separate real  $x_k$  and imaginary  $p_k$  parts of  $\Psi_k$  amplitudes to convert the complex 1<sup>st</sup>-order equation  $i\partial_t \Psi = \mathbf{H}\Psi$ into pairs of <u>real</u> 1<sup>st</sup>-order differential equations.

 $\dot{x}_1 = Ap_1 + Bp_2 - Cx_2 \qquad \dot{p}_1 = -Ax_1 - Bx_2 - Cp_2 \\ \dot{x}_2 = Bp_1 + Dp_2 + Cx_1 \qquad \dot{p}_2 = -Bx_1 - Dx_2 + Cp_1$ 

Then start with classical Hamiltonian. (Designed to give same result.)

$$H_{c} = \frac{A}{2} \left( p_{1}^{2} + x_{1}^{2} \right) + B \left( x_{1}x_{2} + p_{1}p_{2} \right) + C \left( x_{1}p_{2} - x_{2}p_{1} \right) + \frac{D}{2} \left( p_{2}^{2} + x_{2}^{2} \right)$$

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$$\dot{x}_1 = \frac{\partial H_c}{\partial p_1} = Ap_1 + Bp_2 - Cx_2 \qquad \dot{p}_1 = -\frac{\partial H_c}{\partial x_1} = -(Ax_1 + Bx_2 + Cp_2)$$
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$$\dot{p}_{3} = -K \cdot |\mathbf{x}|$$

$$\begin{aligned} \ddot{x}_{1} &= A\dot{p}_{1} + B\dot{p}_{2} - C\dot{x}_{2} \\ &= -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ &= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2} \end{aligned}$$

$$\begin{aligned} \ddot{x}_{2} &= B\dot{p}_{1} + D\dot{p}_{2} + C\dot{x}_{1} \\ &= -B(Ax_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Dx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2}) \\ &= -(AB + BD)x_{1} - (B^{2} + D^{2} + C^{2})x_{2} + C(A + D)p_{1} \end{aligned}$$

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$$= -(A^{2} + B^{2} + C^{2})x_{1} - (A^{2} + B^{2} - Cx_{2}) + C(Ap_{1} + Bp_{2} - Cx_{2})$$

$$= -(A^{2} + B^{2} + C^{2})x_{1} - (A^{2} + B^{2} - A^{2} + B^{2} - A^{2})x_{2} - C(A + D)p_{1}$$

$$\begin{pmatrix} \dot{x}_{1} \\ \dot{x}_{2} \end{pmatrix} = -\left(A^{2} + B^{2} - A^{2} + B^{2} - A^{2} + B^{2} - A^{2} + Cx_{1} + Bx_{2} + Cp_{2} - Cx_{2} + Cx_{1} + Cx_{2} + Cx_{1} + Cx_{2} + Cx_{2} + Cx_{1} + Cx_{1} + Cx_{2} + Cx_{1} + Cx_$$

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Here is an operator view of the QM-Classical connection: Take Schrodinger operator  $i\partial_t = \mathbf{H}$  (with C=0) and square it!

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$$\begin{array}{l} M vs. Classical Equations are identical \\ Equations are identical \\ \vdots \\ z_{2} = \frac{\partial H_{c}}{\partial p_{1}} = Ap_{1} + Bp_{2} - Cx_{2} \\ \dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1} \\ \end{array}$$

$$\begin{array}{l} \dot{p}_{1} = -Ax_{1} - Bx_{2} - Cp_{2} \\ \dot{p}_{2} = -Bx_{1} - Dx_{2} + Cp_{1} \\ \vdots \\ z_{1} = A\dot{p}_{1} \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - B(Bx_{1} + Dx_{2} - Cp_{1}) - C(Bp_{1} + Dp_{2} + Cx_{1}) \\ = -A(Ax_{1} + Bx_{2} + Cp_{2}) - D(Bx_{1} + Dx_{2} - Cp_{1}) + C(Ap_{1} + Bp_{2} - Cx_{2}) \\ = -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2} \\ = -(AB + BD)x_{1} - (B^{2} + D^{2} + C^{2})x_{2} + C(A + D)p_{1} \\ \end{array}$$

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} FOr C - O \\ Is form of 2D Hooke \\ harmonic oscillator \end{pmatrix} = \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_{22} \end{pmatrix} = \begin{pmatrix} K_{11} & K_{22} \\ K_{22} & K_$$

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*Conclusion: 2-state Schro-equation*  $i\hbar \frac{\partial}{\partial t} |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  is like "square-root" of Newton-Hooke.  $\sqrt{|\mathbf{x}\rangle} = -\mathbf{K} \cdot |\mathbf{x}\rangle$ 

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Finally a 2<sup>nd</sup> time derivative (Assume constant A, B, D, and let C=0) gives 2<sup>nd</sup>-order classical Newton-Hooke-like equation:  $|\ddot{x}\rangle = -\mathbf{K} \cdot |\mathbf{x}\rangle$ 

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$$= -(A^{2} + B^{2} + C^{2})x_{1} - (AB + BD)x_{2} - C(A + D)p_{2}$$

$$= -(AB + BD)x_{1} - (B^{2} + D^{2} + C^{2})x_{2} + C(A + D)p_{1}$$

$$\begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} A^2 + B^2 & AB + BD \\ AB + BD & B^2 + D^2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} For \ C = 0 \\ Is \ form \ of \ 2D \ Hooke \\ harmonic \ oscillator \end{pmatrix} = - \begin{pmatrix} \ddot{x}_1 \\ \ddot{x}_2 \end{pmatrix} = - \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\ K_{22} & K_{22} \end{pmatrix} \begin{pmatrix} x_1 & K_{22} \\$$

Here is an operator view of the QM-Classical connection: Take Schrodinger operator  $i\partial_t = \mathbf{H}$  (with  $C \neq 0$ ) and square it!

$$i\frac{\partial}{\partial t} = \begin{pmatrix} A & B-iC \\ B+iC & D \end{pmatrix} \Rightarrow \begin{pmatrix} i\frac{\partial}{\partial t} \end{pmatrix}^2 = \begin{pmatrix} A & B-iC \\ B+iC & D \end{pmatrix}^2 \Rightarrow -\frac{\partial^2}{\partial t^2} = \begin{pmatrix} A^2+B^2+C^2 & AB+BD-i(AC+CD) \\ AB+BD+i(AC+CD) & B^2+D^2+C^2 \end{pmatrix}$$

*Conclusion: 2-state Schro-equation*  $i\hbar \frac{\partial}{\partial t} |\Psi(t)\rangle = \mathbf{H} |\Psi(t)\rangle$  is like "square-root" of Newton-Hooke.  $\sqrt{|\mathbf{x}\rangle} = -\mathbf{K} \cdot |\mathbf{x}\rangle$